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METHOD AND APPARATUS FOR AIR-TO-AIR AIRCRAFT RANGING

TECHNICAL FIELD OF THE INVENTION

This invention relates in general to methods and apparatus for determining the range from one aircraft to another aircraft, and more particularly relates to methods and apparatus for determining the range through use of passive energy detection with directional receivers.

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BACKGROUND OF THE INVENTION

5 The goal of air-to-air passive ranging is to determine the range (distance) from one aircraft, called "ownship", to another, called "target", by means of electromagnetic energy emanating from target. For this purpose, ownship is equipped with directional receivers operable to determine the angle of arrival of the incoming energy at various times during a data collection interval. In most applications, the energy detected will be in the radio frequency (RF) or infrared (IR) portions of the spectrum. By combining the directional measurements with measurements of ownship's position and heading, the range between the two aircraft may be determined. The dominant problem in air-to-air passive ranging is ill-conditioning. Ill-conditioning refers to small errors in the directional measurements causing much larger errors in the range estimate. The sources of ill-conditioning include the limited baseline over which data is measured and the need to infer target motion from the data.

10 Range is estimated from bearing measurements at different locations using some variation of a "triangulation principle". A "baseline" is the distance between two points at which the bearing measurements, called sightings, are taken. Other factors being equal, sightings projected from a long baseline intersect at a greater angle than those taken from a short baseline. Since sightings projected from a short baseline are nearly parallel, a small error in the measured angle results in a

large error in determination of range. Hence, range measurements determined using a long baseline are much more impervious to noise contained in the bearing measurements than are those determined over a short baseline. However, the desire to determine range as quickly as possible restricts the baseline that may be developed during data collection. Furthermore, flight path geometry also dictates the maximum baseline which may be used; for example, if two aircraft are flying towards one another, no baseline will be generated regardless of time.

In air-to-ground ranging, the target can be presumed as being fixed in space. Thus, it can be described by a static model consisting of its coordinates in 3-dimensional space. However, this is not true of the air-to-air problem, where target motion must be considered. The additional degrees-of-freedom implied by the dynamic model must be resolved from the data. This aggravates ill-conditioning because it diminishes the redundancy by which information may be distinguished from noise in the data. Unresolved noise in the data produces ranging errors.

As established by the foregoing, a need has arisen to provide a method and apparatus for measuring the range from one aircraft to another, capable of supplying ranging information quickly, while minimizing the effect of small errors in data measurement on the ranging determination.

SUMMARY OF THE INVENTION

5 In accordance with the present invention, a
air-to-air passive ranging apparatus is provided which
substantially eliminates or prevents the disadvantages and
problems associated with prior art passive ranging
systems.

a 10 In accordance with another aspect of the
invention, a ranging system~~s~~ is provided wherein model
data estimating the flight characteristics of the target
plane are produced and compared with actual flight path
data of the target plane. An error measurement between
the model data and the actual flight path data is
calculated and adjustment is made to model data which
minimizes the error measurement.

15 In yet another aspect of the invention, the
error measurement comprises a means squared sighting
error, a velocity related error, and three position
related errors. A system error is defined as the sum of
the means squared sighting error, the velocity error and
the three position errors. A perturbation model is
20 determined such that the sum of the model data and the
perturbation model data will result in a new set of model
data which reduces the error measurement.

25 In a further aspect of the invention, a method
and apparatus for initializing the model data is provided.
In yet a further aspect of the invention, a method and
apparatus for determining a flight path which minimizes

the effect of noise on the sighting measurements is provided.

BRIEF DESCRIPTION OF THE DRAWINGS

For a more complete understanding of the present invention and the advantages thereof, reference is now made to the following descriptions taken in conjunction with the accompanying drawings in which:

5 FIGURE 1 is a perspective view of an aerial encounter between two aircraft;

 FIGURE 2a-b is a diagram illustrating the ranging problem;

10 FIGURE 3 is a block diagram of the ranging system functional architecture;

 FIGURE 4 is a flow chart describing calculation of the perturbation model;

 FIGURE 5 is a flow chart describing calculation of the perturbation magnitude;

15 FIGURE 6 is a flow chart describing calculation of the initial brackets;

 FIGURE 7 is a flow chart describing a flight path advisor;

20 FIGURE 8 is a diagram describing the formation of the initialization model;

 FIGURE 9 is a graph of the system eigenvalues; and

 FIGURE 10 is a circuit diagram of the ranging system.

DETAILED DESCRIPTION OF THE INVENTION

The preferred embodiment of the present invention is best understood by referring to FIGURES 1-10 of the drawings, like numerals being used for like and corresponding parts of the various drawings.

FIGURE 1 illustrates an aerial encounter between two aircraft. Ownship 10 desires to determine the range between itself and a second aircraft, target 12. Target 12 emits electromagnetic energy, generally indicated by the reference numeral 14. Ownship 10 contains directional receivers 16 operable to determine the angle-of-arrival of the incoming energy forces 14 at various times during a data collection and interval. In addition to the bearing measurements, fixes of ownship's position and heading are monitored by the on-board inertial navigation system (INS) 18. A rectangular coordinate system 19 having an X_1 axis 20, an X_2 axis 22, and an X_3 axis 24, has its center 26 located at ownship. The coordinate system 19 moves in pure translation with ownship 10; however, the directions of the coordinate axes 20-24 are unrelated to the body axes of ownship 10 but are defined in inertial space by the INS 18. For convenience, the bearing measurements from the directional receivers 16 are converted to measurements related to the fixed axes 20-24. The necessary conversions are readily computed from knowledge of ownship's heading as provided by the INS 18.

The location of the target 12 can thus be described as a point in the coordinate system 19 expressed

in terms of the three axes 20, 22, and 24. Alternatively, the target can be described in terms of angles from two of the axes, denoted E (for elevation) and A (for azimuth) and ρ (range). It should be noted that an unconventional definition of elevation has been adopted as being the angle between the X_3 axis 24 and the vector from the center 26 to the target 12. The angle E and A can be computed from the bearing measurements provided by the directional receivers.

The design of the directional receiver 16 depends upon the spectrum of the energy 14 being monitored. For most applications, the energy will be in either the RF (radio frequency) or IR (infrared portions) of the energy spectrum. RF systems are designed to monitor radar transmissions and RF communications from the target 12. An RF system is capable of detecting energy from very long distances, substantially exceeding the distances at which the target could acquire ownship on its own radar. Since ownship 10 is using a passive ranging system, i.e., not transmitting energy signals itself, ownship 10 may be able to locate target 12 without reciprocal detection. This results in a primary advantage of passive ranging, namely, stealth. Furthermore, long acquisition range also affords ownship more time to respond to the target's presence. In a RF system, the directional receiver 16 may be implemented by using a 2-axis RF interferometer, or multiple interferometer units covering different sectors.

5 A directional receiver 16 capable of detecting
infrared energy signals 14 may be designed to respond to
thermally emitted IR radiation from the target. Because
of the shorter wave length IR energy signals, the
direction of the incoming IR energy can be resolved more
precisely. Consequently, the range between the aircraft
can be determined more quickly. The infrared system has a
further advantage in that performance is not linked to
target transmission protocol. However, the maximum
10 acquisition range of the IR system is much shorter than
the RF system, thus allowing the target 12 the advantage
of detecting ownship 10 on its radar or with a comparable
IR system. A directional receiver 16 for the IR system
may be implemented using a FLIR (Forward Looking Infra
15 Red) imager and video tracker.

The angles E and A are subject to measurement by
the directional receiver 16; however, the range must be
determined from the bearing measurements and the INS data.
FIGURES 2a-b illustrate that the range between two
20 aircraft 10 and 12 cannot be determined solely on the
direction of the sighting vector 25, i.e., the angles E
and A, in the absence of acceleration of ownship 10.
FIGURE 2 shows the flight path of ownship 10. The
positions of ownship are shown at t_0 , t_1 , t_2 and t_3 .
25 Vector 28 depicts the distance and direction of the flight
path of ownship 10 between time periods t_0 and t_1 ;
likewise, vector 30 indicates ownship flight path between
 t_1 and t_2 , while vector 32 indicates ownship's flight path

between time periods t_2 and t_3 . Sight line 34 indicates the sight vector between ownship 10 and target 12 at time t_0 ; similarly, sight line 36 indicates the sight vector at time t_1 , sight line 38 indicates the sight vector at time t_2 and sight line 40 indicates the sight vector at time t_3 . Starting point 41 indicates the position of target 12 at the time the first measurement is taken. Vectors 42, 44 and 46 indicate the direction and distance traveled by target 12 during the time period from t_0 to t_3 . From phantom starting point 47, phantom target vectors 48, 50 and 52 indicate a flight path which would provide a possible solution to the range algorithms. From another phantom starting point 53, phantom target vectors 54, 56, and 58 are vectors which would also provide a solution to the ranging algorithm.

At time t_0 , ownship 10 would receive an energy signal from target 12. From this signal, the directional receivers 16 aboard ownship 10 would be able to determine the direction of the sighting vector 25 between ownship 10 and target 12, but would not be able to determine the magnitude (range) of the sighting vector 25. Thus, at time t_0 , ownship 10 can only identify target 12 as being somewhere along sighting line 34. From t_0 to t_1 , ownship flies along vector 28. At time t_1 , ownship takes a another bearing measurement, rendering sight line 36. Once again, ownship 10 knows that target 12 is somewhere along sight line 36, but the exact whereabouts of target 12 is yet unknown. From time t_1 to time t_2 , ownship flies

along vector 30. Since ownship is flying at a constant velocity and direction, vectors 28 and 30 are equal in magnitude and colinear. At time t_2 , another bearing measurement by ownship 10, resulting sight line 38. Whereas the range still cannot be determined from the sight vectors, some pertinent information can be determined from the data.

First, it must be assumed that target 12 is flying at a constant velocity and in a constant direction. This is a valid assumption since most aircraft fly at a nominal speed. Given the assumption of constant velocity, the distance travelled by target 12 between t_0-t_1 and t_1-t_2 must be equal. Also, assuming constant direction, the vectors representing the flight path of target 12 must colinear. Given the sight lines 34-40, as shown in FIGURE 2a, only one set of vectors will meet the criteria of colinearity and equal magnitude for any given point along the sight line 34 at t_0 . For example, given the starting point 41 of target 12 along sight line 34, only the flight path given by vectors 42 and 44 will provide a solution in which the vectors are colinear and equal in magnitude. Likewise, given a starting point 47, only the flight path represented by vectors 48 and 50 will result in a solution having colinear vectors of equal magnitude. Further, given starting point 53 on sight line 34, only the vectors 54 and 56 will provide a suitable solution. Unfortunately, in an actual sighting, the starting point

41 of target 12 will be unknown and must be determined by the data.

Even after another time period, the ambiguity cannot be resolved, as illustrated by the sight path vector 46 of the actual target and vectors 52 and 58 of the phantom target. In order to resolve this ambiguity, ownship must accelerate or decelerate, as shown in FIGURE 2b.

FIGURE 2b depicts the same information as FIGURE 2a, with the exception that ownship 10 decelerated after time period t_2 resulting in a reduced distance travelled over vector 32. Consequently, the direction of sight line 40 taken at time t_3 has changed from FIGURE 2a. Due to the deceleration of ownship 10 at t_2 , without a corresponding deceleration by target 12, only the flight path defined by vectors 42, 44 and 46 meet the criteria of colinearity and equal magnitude, and hence, the ambiguity is resolved. Thus, the magnitude of the sight vector can be determined from the angles of the sight lines 34, 36, 38 and 40.

However, as can be seen from FIGURE 2a, the correct determination of the range is highly dependent upon a accurate measurement of the sight line. For example, a slight change in the angle of sight line 40 taken at t_3 in FIGURE 2b could result in the intersection of the sight line 40 with the endpoint of vector 52, rather than the endpoint of the actual target vector 46. As can be seen in FIGURE 2b, this small error in angle

measurement could result in a miscalculation in range of greater than 20%.

5 The effects of ill-conditioning may be reduced by introducing data redundancy into the equations used to determine the range, i.e., solving the range algorithm over several additional sightings. However, data redundancy may be counterproductive if achieved by reducing the sample interval.

10 The inability of the direct solution method to deal with ill-conditioning has been overcome by the present invention using a method which does not compute range directly, but rather computes the parameters of a kinematic model of the target. Once an acceptable model is created, range may be computed from the model, rather
15 than the actual bearing measurements. Since the model has a fixed number of parameters, redundancy increases monotonically as more data are assimilated.

20 FIGURE 3 illustrates a functional organization of the air-to-air ranging system functional architecture. Input data 60 is stored in a measurement buffer 62 which holds the input data 60 for successive time frames. Data from the measurement buffer 62 is used to create the initialization model 64. Initialization model 64 is
25 available to a buffer containing the resident model 66 from which range estimates 68 are computed. Resident model 66 is also used to generate estimated azimuth and elevation "synthetic measurements" 70. The synthetic measurements 70 are compared with data from the

measurement buffer 62 in a subtracter 72 yielding sighting errors 74. The sighting errors 74 are used in conjunction with the resident model 66 to construct sighting error model 76. A position penalty model 78 is determined using data from the measurement buffer 62, the resident model 66, and data regarding the maximum acquisition range 80. A speed penalty model 82 is assessed using the resident model 66 and data regarding the nominal target speed 84. The position penalty model 78 and the speed penalty model 82 are added to the sighting error model 76 in summation block 86. Data from the summation block 86 along with data from the position penalties 78 and the speed penalty 82 are used to construct an system error model 88. The system error model 88 is used to produce a perturbation model 90. The perturbation model 90 updates the resident model 66. Since a single perturbation of the resident model 66 may not provide a sufficient accuracy between the resident model 66 and the input data 60, the algorithm is iterated until the resident model 66 describes the input data 60 within a predetermined range. After the resident model 66 is sufficiently accurate to correspond with the input data 60, new input data 60 stored in the measurement buffer 62 relating to the next time period is compared with the data from the resident model 66 and the process repeats itself.

In operation, the input data 60 is received from the INS 18 and directional receiver 16. The data includes measurements of time (t), azimuth (A), elevation (E),

ownship position relative to initial starting point (a_1 , a_2 , a_3), and ownship heading (y_1 , y_2 and y_3). Azimuth and elevation data are produced by the directional receivers 16 and are adjusted to conform to the coordinate system 19. Ownship position and heading are produced by the INS 18 and likewise conform to the coordinate system 19. Timing data is received from an internal clock.

The input data 60 is stored in a measurement buffer 62. While the input data can be processed substantially in real-time, the measurement buffer 60 is necessary to store the data for use in different tasks, such as producing the initialization model 64, the sighting error 74 and the position penalty 78.

The initialization model block 64 computes the initial estimate of the model parameters. The six model parameters, denoted as the vector \underline{m} , describe the position of the target 12 as a function of time, including three coordinates of position (x_1 , x_2 and x_3) and a three dimensional velocity vector having components v_1 , v_2 and v_3 . Since target 12 is assumed to be flying at a constant speed and direction, the velocity \underline{v} can be assumed constant. Calculation of the initialization model block 64 will be described in greater detail in conjunction with FIGURES 8 and 9.

Once the initial model 64 has been determined, it is used as the first resident model 66. The resident model 66 is the system's current best estimate of the target location and velocity. The ranging algorithm

revises the resident model 66 at each time interval to correspond with the actual input data over a period of time. Once the resident model 66 can no longer be improved, the system's best estimate of range can be obtained from the model parameters. As each new set of input data 60 is received, a perturbation model 90 is formed to update the resident model 66. The computation of the perturbation model is discussed in connection with FIGURES 3-6.

Synthetic measurements 70 are measurements of elevation (E) and azimuth (A) calculated from the data from the resident model 66. The synthetic measurements 70 would correspond to the actual input data 60 if the resident model 66 was an accurate description of the target 12 and the input data 60 was accurate and noise-free. The relationship between the azimuth and elevation and the model parameters are as follows:

$$A_{mi} = \tan^{-1} \frac{x_2(0) + t_i v_2 + a_2(t_i)}{x_1(0) + t_i v_1 + a_1(t_i)}$$

$$E_{mi} = \tan^{-1} \frac{(x_1(0) + t_i v_1 + a_1(t_i))^2 + (x_2(0) + t_i v_2 + a_2(t_i))^2}{x_3(0) + t_i v_3 + a_3(t_i)}$$

where E_i and A_i are the values computed from the resident model 66 at time t_i .

An error function $J(\underline{m})$ is defined by the equation:

$$J(\underline{m}) = 1/N \sum_{i=0}^{N-1} (E_{oi} - E_{mi})^2 + (A_{oi} - A_{mi})^2$$

where E_{oi} and A_{oi} are the observed elevation and azimuth at time t_i and E_{mi} and A_{mi} are the values computed from the resident model 66 at time t_i . $J(\underline{m})$ is the mean squared sighting error evaluated with the data from the resident model, \underline{m} . The object of the ranging system is to determine a model which minimizes $J(\underline{m})$. Once this model is obtained, range is readily computed from the model parameters. To obtain the model which minimizes $J(\underline{m})$, the resident model 66 is "perturbed" in a direction and magnitude which produces a new resident model which improves correlation between the actual observed data stored in the measurement buffer 62 and the synthetic measurements 70 calculated from the model.

The subtractor 72 provides the sighting errors 74, $E_{oi} - E_{mi}$ and $A_{oi} - A_{mi}$, used in the calculation of the mean squared error function. The mean squared sighting error along with the sighting error perturbation model 76.

The mean squared sighting error of the perturbed model is denoted:

$J' = J(\underline{m} + \delta \underline{m})$, where $\delta \underline{m}$ is the perturbation.

This equation (the "perturbation equation") may be approximated by the second degree model:

$$J' = J(\underline{m}) + \delta \underline{m}^T \underline{g} + (1/2) \delta \underline{m}^T \underline{H} \delta \underline{m}$$

where \underline{g} is the gradient of $J(\underline{m})$, a vector whose elements are the partial derivatives of $J(\underline{m})$ with respect to the model parameters, i.e.:

$$g_r = (\partial J / \partial m_r), \text{ for } r = 1, 2, \dots, 6$$

The vector \underline{g} can also be described in relation to the sighting error as:

$$\underline{g}_J = (-2/N) \sum_{i=0}^{N-1} (\epsilon_i \nabla E_{mi} + \alpha_i \nabla A_{mi})$$

where ϵ_i and α_i are the instantaneous sighting errors for elevation and azimuth:

$$\epsilon_i = E_{oi} - E_{mi} \text{ and } \alpha_i = A_{oi} - A_{mi}$$

and ∇E_{mi} and ∇A_{mi} are the gradients of E_{mi} and A_{mi} with respect to the parameters of the model \underline{m} .

The "Hessian" matrix, H , is denoted as:

$$H = [h_{rs}] = [(\partial^2 J) / (\partial m_r \partial m_s)]$$

where m_r and m_s respectively refer to the r^{th} and s^{th} parameter in the set of six model parameters. The H

matrix for the sighting error may be approximated by the equation:

$$H_J = (2/N) \sum_{i=0}^{N-1} (\nabla E_{mi} \nabla E_{mi}^T + \nabla A_{mi} \nabla A_{mi}^T)$$

Since the vectors involving ∇E_{mi} and ∇A_{mi} have already been computed in determining the gradient of $J(\underline{m})$ (denoted as \underline{g} in the equations), no new differentiations are required to compute the H matrix.

The Hessian matrix, H_J , and the gradient of $J(\underline{m})$, \underline{g}_J , of the sighting error perturbation equation $J(\underline{m})$ are summed with corresponding Hessian matrices and gradients found in relation to speed and position penalties 78 and 82, described below, to render a system error model 88. The system error model 88 is used to create a perturbation model 90.

Position penalties model 78 and speed penalty model 82 provide constraints to the ranging algorithm. "Constraints" provide a mechanism for including real-world knowledge which is not explicitly present in the input data 60 into the ranging algorithms. Constraints restrict the domain of feasible solutions to target models with realistic parameters. Four constraints are employed in the system: a speed constraint, and three position constraints. The position constraints include a maximum range constraint, an elevation position constraint, and an

azimuth position constraint. The use of constraints speeds convergence of the resident model 66 and reduces the effect of ill conditioning. While the use of constraints requires additional computational costs, the costs are offset by reduction in the iterations required to achieve convergence of the resident model 66.

The actions of the constraints are indirect rather than direct, since the constraints control functions of the model parameters rather than the parameters themselves. For instance, the speed constraint acts upon the magnitude of the target velocity, i.e., $(v_1^2 + v_2^2 + v_3^2)^{1/2}$. Clearly, the speed may remain constant even though two or more of the velocity parameters change. Also, it is clear that changes in the position parameters, x_1 , x_2 and x_3 do not effect the speed of the target. Therefore, all target parameters may change without affecting the speed, and hence, the speed constraint.

The speed constraint imposes a penalty when the velocity parameters, v_1 , v_2 and v_3 , describe a target which has a magnitude of velocity deviating significantly from a nominal speed. Most aircraft of the same class fly at approximately the same speed, thus a predetermined nominal speed may be set, and a speed constrained perturbation model may be determined that requires that the target be within a specified tolerance of the predetermined nominal speed.

The function used to determine the speed penalty is:

$$\psi = K[(\mu - \mu_R)/\Delta\mu]^{2M}$$

where: μ = target speed determined from model
 μ_R = nominal (predetermined) speed

K = weight of the constraint
 $\Delta\mu$ = bandwidth of the constraint
 M = order of the constraint

All of the elements of the speed penalty function are non-negative numbers. The element " μ " is usually determined as the magnitude of the velocity factor, while μ_R is chosen as the nominal speed of a particular aircraft. The bandwidth, $\Delta\mu$, is chosen in proportion to the uncertainty with which the true target speed is known. The weight = K , is chosen equal to the desired value of the speed penalty when model speed deviates from nominal speed by one bandwidth. The order, M , increases the influence of the constraint beyond the bandwidth. In the preferred embodiment, orders above $M = 8$ are not required, and are avoided, because of the associated numerical overflow and underflow problems in the evaluation of the derivatives in the stop band.

The first and second order partial derivatives of ψ are computed as constituents of the perturbation

equations described in conjunction with the sighting error
 perturbation model $J(\underline{m})$ to render a corresponding Hessian
 matrix H_ψ and a gradient vector \underline{g}_ψ . Since the constraint
 is entirely velocity-related, the arguments are reduced
 from the six used in the sighting error perturbation
 equations, to three used in the speed constrained
 perturbation equations. Thus, in computing in the Hessian
 matrix of the speed constrained perturbation equations,
 only six second order partial derivatives are computed,
 rather than the twenty-one computed in connection with the
 sighting error perturbation equation. Furthermore, the
 constraint ψ is not time-dependent, as is the calculation
 for mean squared sighting error $J(\underline{m})$. Therefore,
 time-iterative calculations need not be performed to
 compute ψ .

The position constraints require that the
 components of the target model associated with the initial
 position lie within a "cone" centered about the initial
 sighting vector. The angular cross-section of the cone is
 sized according to the uncertainty of sighting
 measurements, and is a statistical choice.

A penalty is associated with each position
 constraints: a elevation position penalty, an azimuth
 position penalty, and a maximum range penalty. The
 equations for these penalties are:

$$P_e = K_e [(E_{oi} - E_{mi}) / \Delta E]^{2Me}$$

$$P_a = K_e [(A_{oi} - A_{mi})/\Delta E]^{2Ma}$$

$$P_r = K_r \exp [\alpha(\rho - \rho_r)/\rho_r]$$

The constants K_e , K_a , K_r , Me , Ma , and α are chosen as described in conjunction with the speed penalty. ΔE and ΔA are chosen in proportion to the uncertainty with which the true target elevation and azimuth are known. The factor ρ_r is set equal to the maximum acquisition range.

Three position penalty models are constructed, one for each of the three penalty equations. The position penalty models are determined as described above in connection with the sighting error model rendering equations for a Hessian matrix and gradient vector for each penalty model. As with the velocity penalty, the Hessian matrix for the position penalty equations is limited to the three position parameters - x_1 , x_2 and x_3 .

After computation of the sighting error model 76, speed constrained model 82 and three position models 78, the models are summed to create a system model 88. The equations used to define the system model are:

$$J_s = J(\underline{m}) + \psi + P_e + P_a + P_r$$

$$H_s = H_j + H_\psi + H_{Pe} + H_{Pa} + H_{Pr}$$

$$g_s = g_j + g_\psi + g_{Pe} + g_{Pa} + g_{Pr}$$

Several methods exist for minimizing the mean squared error $J_s(\underline{m})$. The preferred method is the

Newton-Raphson method, which converges faster and provides for larger perturbations than other methods, such as the gradient method. A key step of the Newton-Raphson method is to solve $H_s \delta \underline{m} = \underline{g}_s$. A flow chart of the preferred method for solving this equation is given in FIGURE 4. In block 92, a Cholesky factorization is performed. The Cholesky factorization involves factoring the H_s matrix into lower and upper diagonal matrices, L and L^T . In block 94, the equation $Ly = \underline{g}_s$ is solved for the vector y (L and \underline{g}_s are known). In block 96, the equation, $L^T \delta \underline{m} = y$ is solved for $\delta \underline{m}$.

The accuracy of the perturbation model, $\delta \underline{m}$, computed in this manner, may be improved as illustrated in blocks 98-107 of FIGURE 4. In block 98, the solution for $\delta \underline{m}$ from block 96 is set as the initial solution, $\delta \underline{m}_0$. In block 100, a residual, \underline{r}_0 , is determined from the equation $\underline{r}_0 = \underline{g} - L L^T \delta \underline{m}_0$, using a double precision computer arithmetic. In block 102, the equation $L L^T \underline{y}_0 = \underline{r}_0$ is solved for the "correction", \underline{y}_0 . In block 104, \underline{y}_0 is added to $\delta \underline{m}_0$ to provide a more accurate solution for the perturbation, $\delta \underline{m}$. As indicated in decision block 106, if more accuracy is desired the loop starting at block 98 may be repeated. This algorithm used to improved accuracy of the initial perturbation solution has the advantage requiring only one double precision calculation for the remainder in block 100; the other two calculations in blocks 102 and 104 may be performed using single precision computations.

5 The system error model has both direction and
 magnitude. However, to establish the optimum magnitude of
 the perturbation, the preferred method is to perform a
 linear search, which allows non-linearities of higher
 degree in the system error function to influence the
 optimum magnitude. The purpose of the linear search is to
 determine the weight of the perturbation δ_m at which the
 system error J_s is minimized. In the preferred embodiment
 10 Davidon's cubic interpolation method is utilized.
 Davidon's cubic interpolation method is described in
 detail in "Methods of Optimization", G.R. Walsh, John
 Wiley & Sons (1975).

15 A flow chart outlining the Davidon algorithm is
 illustrated in FIGURE 5. In block 108, a "bracket" having
 an upper and lower boundaries is established. Within the
 bracket, the optimum weight (magnitude) of the
 perturbation is located. The bracket may be determined by
 a trial-and-trial procedure, or by the procedure discussed
 hereinafter in connection with FIGURE 6.

20 Block 110 is the first step of iterative loop
 used to narrow the boundaries of the bracket. In block
 110, the width of the bracket, i.e., the distance between
 the upper and lower boundaries, is determined. In block
 112, the width of the bracket is compared to a
 25 predetermined value. The predetermined value corresponds
 to a width at which further precision in determining the
 optimum weight of the perturbation no longer warrants
 additional computational time. If the width of the

bracket is less than or equal to the predetermined value, then the optimum weight is set to a value equal to the arithmetic average of the boundary weights as illustrated in block 114. If the widths of the bracket is greater than the predetermined value in block 112, then Davidon's cubic interpolation algorithm is used to estimate the location of the optimum weight within the bracket in block 118. Davidon's cubic interpolation method computes a cubic polynomial estimating the system error J_s based on the values of the system error and the derivative of the system error at the upper and lower bracket boundaries. The minimum value of the cubic polynomial can be determined from its derivative. The estimated optimum weight is set to the point at which the cubic polynomial is at a minimum. In block 120, the actual system error and its derivatives in the direction of the perturbation are computed for the estimated optimum weight determined in block 118. In decision block 122, if the derivative determined in block 120 is less than zero, then the lower boundary of the bracket moves to the location of the estimated optimum weight determined in block 118, thus narrowing the width of the bracket. If the derivative is not less than zero in decision block 122, and the derivative is greater than zero in decision block 126, the upper boundary of the bracket moves to the location of the estimated optimum weight, as described in block 130. If the derivative is either less than zero in block 122 or greater than zero in block 126, then the loop repeated

starting at block 110. If the derivative is equal to zero, then the optimum weight is set equal to the estimated optimum weight.

5 After a number of iterations, the width of the bracket will be less than the predetermined value given in block 110. At this point, the linear search is completed, and the optimum weight is selected as the average between the boundary weights in block 114, or as specified in block 128.

10 *a* In some instances, J_s may be substantially flat around its minimum, resulting in a slow convergence. If successive iterations of the bracketing algorithm are producing new boundaries for one side of the bracket without alternation between upper and lower boundaries, then steps must be taken to aid convergence. For example, 15 if successive iterations of the algorithm have produced only incremental changes in the upper boundary, then the lower boundary should be moved closer to the upper boundary by a predetermined factor, such as by one third 20 of the width between the upper and lower boundaries.

25 FIGURE 6 illustrates a flow chart for determining the initial bracket boundaries. In block 132, the system error and its derivative along the direction of the perturbation is computed for a weight equal to zero. In decision block 134, the derivative of the system error is checked for a negative condition; if the derivative is equal to or greater than zero, then an error flag is returned in block 136, indicating that no reduction of

system error is possible in the direction of the
 perturbation as calculated. If the derivative is less
 than zero, then the lower boundary is set for a weight of
 zero in block 138. Assuming a non-negative derivative of
 5 the system error, a first weight (W_1) of the model
 perturbation is determined such that the larger of the
 velocity penalty or the position penalty will equal the
 system error determined in block 132. In block 142, the
 upper boundary is set to equal W_1 . In block 144, the
 10 derivative of the system error along the direction of the
 perturbation at the upper boundary is evaluated for use in
 the Davidon's algorithm of FIGURE 5. If the derivative of
 the system error is negative at W_1 , Davidon's cubic
 polynomial will not be able to compute a suitable
 15 approximation of the system error between the upper and
 lower boundaries. In this instance, the boundaries are
 relocated as described in block 145. If the value of the
 system error at W_1 exceeds or equals the value of the
 system error at the lower boundary, then the upper
 20 boundary is set at $1/3 W_1$. Otherwise, the lower boundary
 is set at W_1 and the upper boundary is set at $2W_1$.

After the optimum weight of the perturbation is
 determined, the resident model 66 is updated by adding the
 perturbation multiplied by the optimum weight to the
 25 resident model 66. The new resident model is tested to
 determine whether further iterations would be fruitful,
 i.e., whether successive perturbations of the resident
 model will have a significant effect on the range

determination. By calculating a magnitude of the perturbation, and comparing it with a predetermined limit, the capacity for further improvement of the resident model 66 is evaluated. If no further iterations are necessary, the resident model is considered valid and can be used to compute the range at any time in the data collection interval. The system will continue iterate after the range determinations are valid in order to update the range information and to detect path deviations by target 12.

As described in connection with FIGURES 2a-b, it is necessary ownship 10 to accelerate relative to target 12, in order to allow computation of the range. A method for determining the "optimum maneuver" which will produce acceleration (or deceleration) in a direction having maximum insensitivity to errors in the measured sighting vectors 25 is illustrated in the flow chart of FIGURE 7. In Block 146, the sighting vector (\underline{w}) 25 is shown in terms of the elevation (E) and the azimuth (A). In Block 148, a matrix W is defined as the sum of the sighting vector times its transpose over N sightings. The eigenvalues of the matrix W are determined in Block 150. In block 152, an eigenvector, \underline{v} , associated with the smallest eigenvalue found in Block 150, is determined. This eigenvector is used in Block 154 to advise ownship 10 of the change of direction which will minimize errors due to noise in the sighting data. Hence, if ownship 10 is travelling with a normalized velocity vector of (1,0,0) and the

eigenvector, \underline{u} , has a value of $(0,1,0)$, the new direction would be equalled to $\underline{v} + k\underline{u} = (1,k,0)$, where k is a positive or negative real value. The factor k is a predetermined value chosen to reflect ownship's ability to change directions substantially instantaneously. The eigenvalues and eigenvectors of matrix W may be determined using well known mathematical techniques.

As previously shown, range cannot be determined when both target 12 and ownship 10 fly with constant velocity. However, range can be estimated by assuming target speed (the magnitude of the target's velocity). By determining the estimated range, all of the model parameters are established. Thus, the initialization model 64 may be established after an estimated range is computed. The preferred method of computing the initialization model 64 is illustrated in FIGURES 8 and 9. The method uses the results obtained in the flight path advisor illustrated in FIGURE 7.

FIGURE 8 illustrates relative target positions at time t_0 and time t_i . The axis \underline{u}_1 and \underline{u}_2 are eigenvectors of the matrix W , defined in relation to FIGURE 7, which correspond to the largest and second largest eigenvalues respectively. Typical eigenvalues of the matrix W are illustrated in FIGURE 9.

To solve for the initial value of the target velocity, \underline{v}_T :

$$\underline{v}_T = [\langle \underline{u}_1, \underline{v}_0 \rangle - \rho_0 \beta \cos((\theta_D/2) + \phi)] \underline{u}_1 + \\ [\langle \underline{u}_2, \underline{v}_0 \rangle + \rho_0 \beta \sin((\theta_D/2) + \phi)] \underline{u}_2 + \\ [\langle \underline{u}_3, \underline{v}_0 \rangle] \underline{u}_3$$

where:

$\underline{u}_1, \underline{u}_2, \underline{u}_3$ - eigenvectors corresponding to largest, second largest, and third largest eigenvalues of matrix W

\underline{v}_0 - ownship velocity vector

θ_D - angle between θ_0 and θ_i , calculated as

$(12\lambda_2)^{1/2}$, where λ_2 is the second largest eigenvalue of matrix W.

ϕ - angle between \underline{w}_0 and \underline{v}_r , where \underline{v}_r is the relative velocity vector

The acquisition range is given by:

$$\rho_0 = [-B - (B^2 - AC)^{1/2}] / A$$

where $A = \beta^2$

$$B = \beta [\langle \underline{u}_2, \underline{v}_0 \rangle \sin((\theta_D/2) + \phi) - \\ \langle \underline{u}_1, \underline{v}_0 \rangle \cos((\theta_D/2) + \phi)]$$

$$C = \mu_0^2 - \mu_T^2 \quad (\mu_0 \text{ and } \mu_T \text{ are the speeds of the ownship and target})$$

$$\text{and } \beta = [(f^2 + s_1^2)/(t_1^2(1 + s_1^2))]^{1/2}$$

$$\text{where } S_1 = \tan(\theta_1); S_2 = \tan(\theta_2)$$

$$f^2 = [(t_1/t_2)^2 (1 + s_1^2)^2 - 2(t_1/t_2) (1+s_1^2) (s_1^2+(s_1/s_2)) + (s_1^2+(s_1/s_2))^2] s_2^2/(s_2-s_1)^2$$

the initial target position, \underline{c} in FIGURE 9, can be determined as:

$$\underline{c} = p_0 (\cos((\theta_D/2)\underline{u}_1 - \sin(\theta_D/2)\underline{u}_2)$$

Referring now to FIGURE 10, a hardware implementation of the ranging system is illustrated. A processor 180 is connected to a program memory 181 and a storage memory 182 and two arithmetic coprocessors 184 and 186. Input data measurements 60 are input to the processor 180, and range determination outputs are output from the processor 180.

Comparing FIGURE 3 with the hardware implementation of FIGURE 10, the storage memory 182 is used to implement the measurement buffer 62 and to store interim calculations of the synthetic measurements 70, sighting errors 74, sighting error model 76, position penalty model 78, speed penalty model 82, system error

model 88, perturbation model 90, resident model 66, and initialization model 64. The processor 180 manages data transfers between the storage memory 182, the arithmetic coprocessors 184 and 186, and the data input and output 60 and 188. The program memory 181 supplies the instruction for processor 180. The program memory 181 may be internal to the processor 180. In the preferred embodiment, a TMS 320 signal processor is used as the processor 180, although other microprocessors may be substituted as well. The arithmetic coprocessors 184 and 186 are used to facilitate calculation of the Hessian matrices and the gradient vectors associated with sighting error model 76, the position penalty model 78, and the speed penalty model 82. Also, the arithmetic coprocessors 184 and 186 are used to determine the magnitude and direction of the perturbation model and to calculate the initialization model 64.

Input data measurements 60 are preferably filtered prior to input to the memory 182. The filter serves to enhance the signal to noise ratio of the data, while providing a down-sampling function.

It is important to note that other implementations for the ranging system may be substituted for the implementation illustrated in FIGURE 10. For example, a plurality of processors 180 may be utilized in order to increase the processing speed of the implementation. Since increased processing speed may allow the input data to be received at shorter time

intervals, this alternative may be desirable for some applications. If multiple processors are used, it may be advantageous to provide separate memories for some or all of the processors in order to prevent a bottleneck from occurring at the storage memory 182. Alternatively, if greater processing speed is required, the ranging system could be implemented using a pipeline configuration, wherein discrete arithmetic components replaced the functions performed by the processor 180. However, while a pipeline configuration can greatly increase processing speed, the number of additional components used in this implementation may result in a system which is too large or too costly in comparison with the benefits of increased processing speed.

Although the present invention has been described in detail, it should be understood that various changes, substitution and alterations can be made therein without departing from the spirit and scope of the invention as defined by the appended claims.

TECHNICAL ADVANTAGES OF THE INVENTION

It is an important technical advantage of the invention that the range between two aircraft can be accurately calculated from data subject to noise.

5 It is a further technical advantage that the invention can determine the range between two aircraft from passively detected data.

It is another technical advantage that a flight path optimizing ranging performance may be calculated.

10 It is yet another technical advantage that the invention supplies ranging information quickly.

It is yet a further technical advantage that the invention provides range information from directional data.

15 It is yet a further technical advantage that the invention provides a method for determining initial estimates of the targets position and velocity.